

EPAs ToxCast Research Program: Developing Predictive Bioactivity Signatures for Chemicals

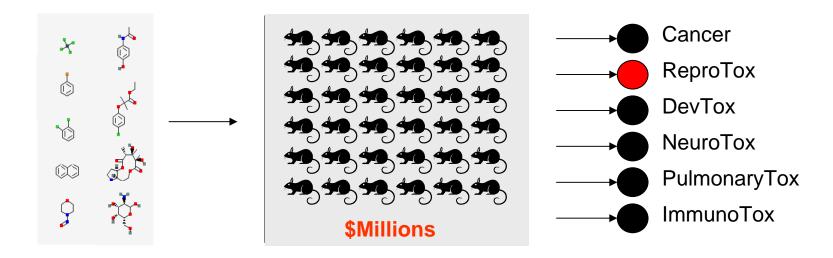
HESI SOT Luncheon Seattle Washington

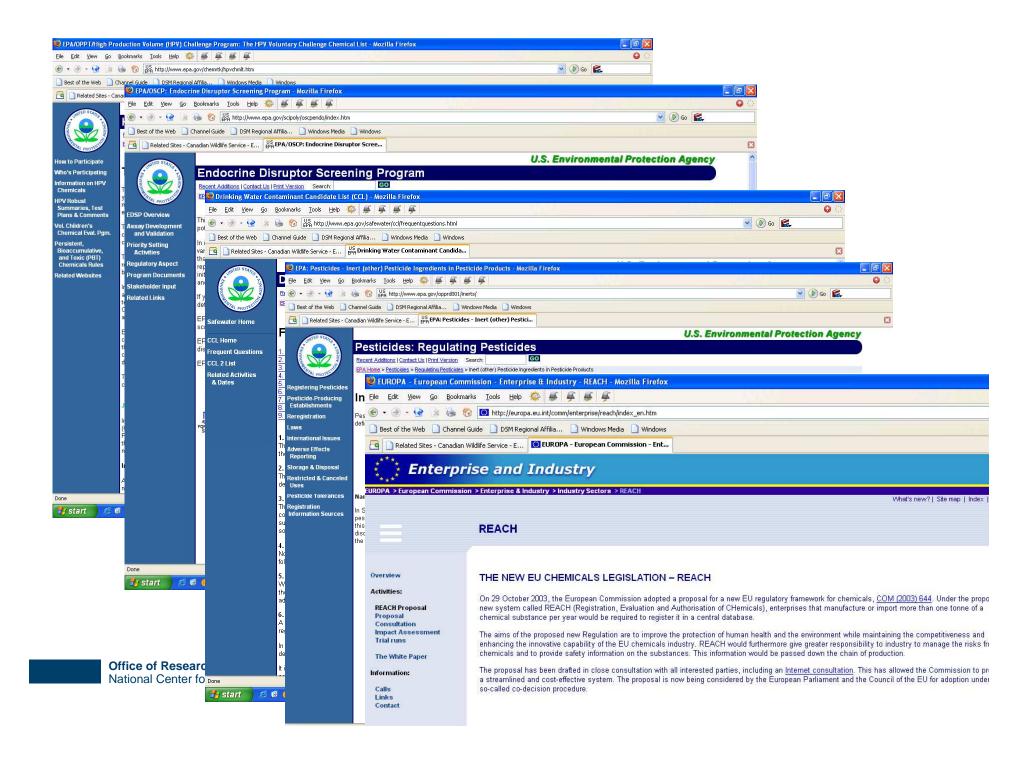




Current Approach for Toxicity Testing

in vivo testing



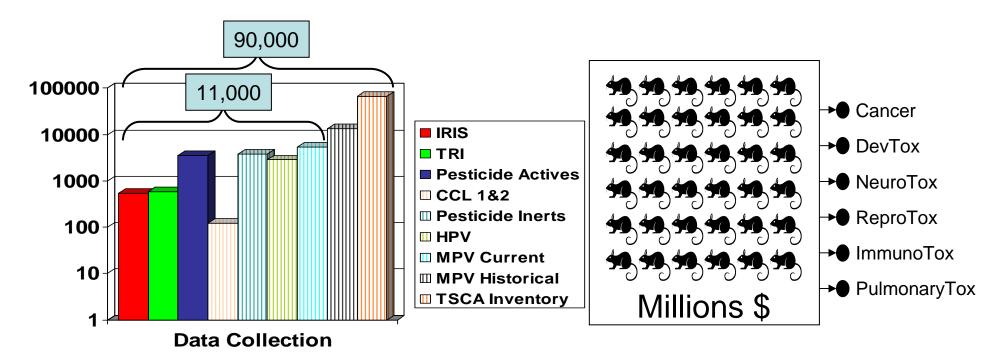




Putting Numbers on the Problem

Too Many Chemicals

Too High a Cost



...and not enough data.





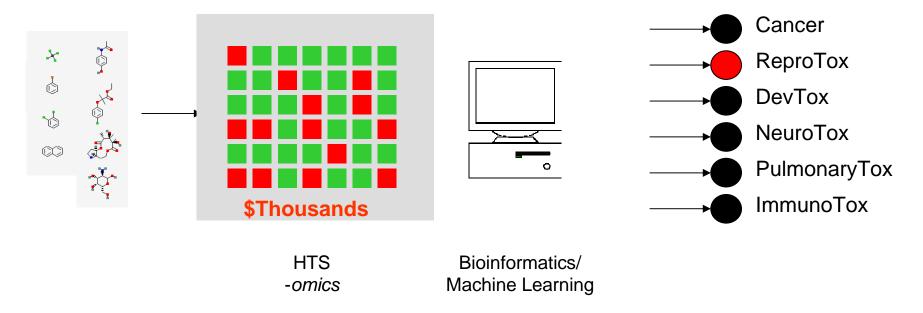
"...to integrate modern computing and information technology with molecular biology to improve Agency prioritization of data requirements and risk assessment of chemicals"

www.epa.gov/ncct



Future of Toxicity Testing

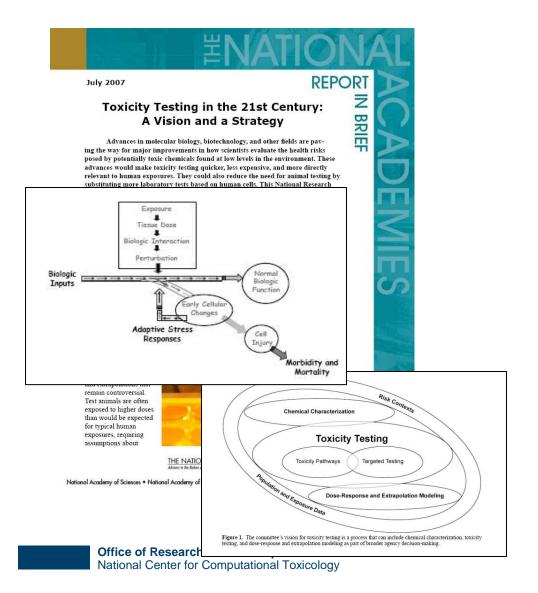
in vitro testing in silico analysis



EPAs Approach: The ToxCast Research Program



Transforming Toxicology



POLICYFORUM

Transforming Environmental Health Protection

Francis S. Collins,1*1 George M. Gray,2* John R. Bucher3*

n 2005, the U.S. Environmental Protection throughput screening (HTS) and other auto-Agency (EPA), with support from the U.S. mated screening assays into its testing ate high false-negative rates. In contrast, in National Toxicology Program (NTP), program. In 2005, the EPA established the the EPA, NCGC, and NTP combined effort, funded a project at the National Research Council (NRC) to develop a long-range vision for toxicity testing and a strategic plan for NTP and EPA, with the NCGC, are promotimplementing that vision. Both agencies wanted future toxicity testing and assessment paradigms to meet evolving regulatory needs. Challenges include the large numbers of substances that need to be tested and how to incorporate recent advances in molecular toxicology, computational sciences, and information Toxicity pathways. In vitro and in vivo tools are being used to identify cellular technology; to rely increasingly on human as opposed to animal data; and to offer increased efficiency in design and costs (1-5). In response, the NRC Committee on Toxicity responses after chemical exposure expected to result in adverse health effects (7). HTS Testing and Assessment of Environmental for drug development, and screening of >100,000 compounds per day is routine (8). Agents produced two reports that reviewed current toxicity testing, identified key issues, and developed a vision and implementation strategy to create a major shift in the assessment of chemical hazard and risk (6, 7). Although the NRC reports have laid out a solid theoretical rationale, comprehensive and rigorously gathered data (and comparisons with historical animal data) will determine whether the hypothesized improvements will be realized in practice. For this purpose, NTP, EPA, and the National Institutes of Health Chemical Genomics Center (NCGC) (organizations with expertise in experimental toxicology. computational toxicology, and high-throughput technologies, respectively) have established a collaborative research program.

EPA, NCGC, and NTP Joint Activities

In 2004, the NTP released its vision and roadmap for the 21st century (1), which established initiatives to integrate high-

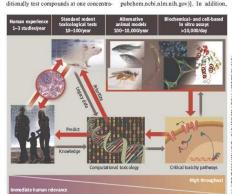
³Director, National Human Genome Research Institute (NHGRI), National Institutes of Health, Bethesda, MD 20892; ²Assistant Administrator for the Office of Research and Development, U.S. Environmental Protection Agency, Washington, DC 20460; ³Associate Director, U.S. National Toxicology Program, National Institute of Environmental Health Sciences (NIEHS), Research Triangle Park, NC

*The views expressed here are those of the individual authors and do not necessarily reflect the views and policies of their respective agencies.

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We propose a shift from primarily in vivo animal studies to in vitro assays, in vivo assays with ower organisms, and computational modeling for toxicity assessments.

tion, usually between 2 and 10 µM, and toler-National Center for Computational Toxi- all compounds are tested at as many as 15 concentrations, generally ranging from ~5 nM to ~100 μM, to generate a concentration response curve (9). This approach is highly reproducible, produces significantly lower level of disease-specific models in vivo to a predominantly predictive science focused the traditional HTS methods (9), and facilion broad inclusion of target-specific, mechanism-based, biological observations in informatics platform has been built to compare results among HTS screens; this is being expanded to allow comparisons with historical toxicologic NTP and EPA data (http://ncgc.nih.gov/pub/openhts). HTS data collected by EPA and NTP, as well as by the NCGC and other Molecular Libraries Initiative centers (http://mli.nih.gov/), are being made publicly available through Webbased databases [e.g., PubChem (http:// pubchem.ncbi.nlm.nih.gov)]. In addition.



Transforming toxicology. The studies we propose will test whether high-throughput and computational toxicology approaches can yield data predictive of results from animal toxicity studies, will allow prioritization of chemicals for further testing, and can assist in prediction of risk to humans.

15 FEBRUARY 2008 VOL 319 SCIENCE www.sciencemag.org

cology (NCCT). Through these initiatives,

ing the evolution of toxicology from a pre-

dominantly observational science at the

methods are a primary means of discovery

However, drug-discovery HTS methods tra-

vitro (1, 4) (see figure, below).

Science, Feb 15, 2008





U.S. scientists are taking the first step towards testing potentially hazardous chemicals on cells grown in a laboratory, without using live animals.

Two government agencies are looking into the merits of using high-speed

automated robots to carry out tests.

time and number of animals used in

household chemicals.

effects.

Boston.

Faster and cheaper

rely less on animal studies.

The long-term goal is to reduce the cost,

screening everything from pesticides to

The move follows calls for scientists to

hundreds of thousands of chemical tests

a day to identify chemicals with toxic

Details were published in the journal Science and discussed at the annual

meeting of the American Association for

the Advancement of Science (AAAS) in

Speaking in a live link-up, Dr. Francis Collins, Director of the National Human

National Institute of Health (NIH), said

Genome Research Institute at the

Robots would be able to carry out



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Phased Development of ToxCast

Phase	Number of Chemicals	Chemical Criteria	Purpose	Number of Assays	Cost per Chemical	Target Date
I	320	Data Rich (pesticides)	Signature Development	>400	\$20k	FY07-08
lla	>300	Data Rich Chemicals	Validation	>400	\$15-20k	FY09
llb	>100	Known Human Toxicants	Extrapolation	>400	\$15-20k	FY09
llc	>300	Expanded Structure and Use Diversity	Extension	>400	\$15-20k	FY10
III	Thousands	Data poor	Prediction and Prioritization	???	\$10-15k	FY11-12

- ➤ Affordable science-based system for categorizing chemicals
- ➤ Increasing confidence as database grows
- ➤ Identifies potential mechanisms of action
- > Refines and reduces animal use for hazard ID and risk assessment

National Center for Computational Toxicology

ToxCast™ Program

Predicting Hazard, Characterizing Toxicity Pathways, and Prioritizing the Toxicity Testing of Environmental Chemicals

Introduction

In 2007, EPA launched ToxCast™ in order to develop a cost-effective approach for prioritizing the toxicity testing of large numbers of chemicals in a short period of time. Using data from state-of-the-art high throughput screening (HTS) bioassays developed in the pharmaceutical industry, ToxCast™ is building computational models to forecast the potential human toxicity of chemicals. These hazard predictions will provide EPA regulatory programs with science-based information helpful in prioritizing chemicals for more detailed toxicological evaluations, and lead to more efficient use of animal testing.

In its first phase, ToxCast™ is profiling over 300 well-characterized chemicals (primarily pesticides) in over 400 HTS endpoints. These endpoints include biochemical assays of protein function, cell-based transcriptional reporter assays, multi-cell interaction assays, transcriptomics on primary cell cultures, and developmental assays in zebrafish embryos. Almost all of the compounds being examined in Phase 1 of ToxCast™ have been tested in traditional toxicology tests, including developmental toxicity, multi-generation studies, and subchronic and chronic rodent bioassays. ToxRefDB, a relational database being created to house this information, will contain nearly \$1B worth of toxicity studies in animals when completed. ToxRefDB is integrated into a more comprehensive data management system developed by NCCT called ACToR (Aggregated Computational Toxicology Resource), that manages the large-scale datasets of ToxCast™.

developed by NCCT called ACToR (Aggregated Computational Toxicology Resource), that manages the large-scale datasets of ToxCast™.

ACToR is comprised of several independent data repositories linked to a common database of chemical structures and properties, and to tools for development of predictive HTS and genomic bioactivity signatures that strongly correlate with specific toxicity endpoints from ToxRefDB. These ToxCast™ signatures will be defined and evaluated by their ability to predict outcomes from existing mammalian toxicity testing, and identify toxicity pathways that are relevant to human health effects.

The second phase of $ToxCast^{m}$ will screen additional compounds representing broader chemical structure and use classes, in order to evaluate the predictive bioactivity signatures developed in Phase I. Following successful conclusion of Phases I and II, $ToxCast^{m}$ will provide EPA regulatory programs an efficient tool for rapidly and efficiently screening compounds and prioritizing further toxicity testing.

ToxCast™ Navigation

Introduction

ToxCast™ Chemicals

ToxCast™ Assays

ToxCast™ Information

Management

ToxCast™ Partnerships

ToxCast™ Contractors

ToxCast™ Presentations

ToxCast™ Publications

ToxCast™ News



Key Components of a Proof of Concept

- Assays covering Toxicity Pathways
 - Chemicals
- Linkage to Traditional Phenotype Findings
 - Data Analysis and Interpretation



The ToxCast Team



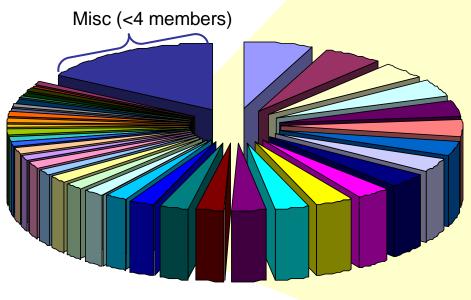


The Chemicals



Chemical Classes in ToxCast_320

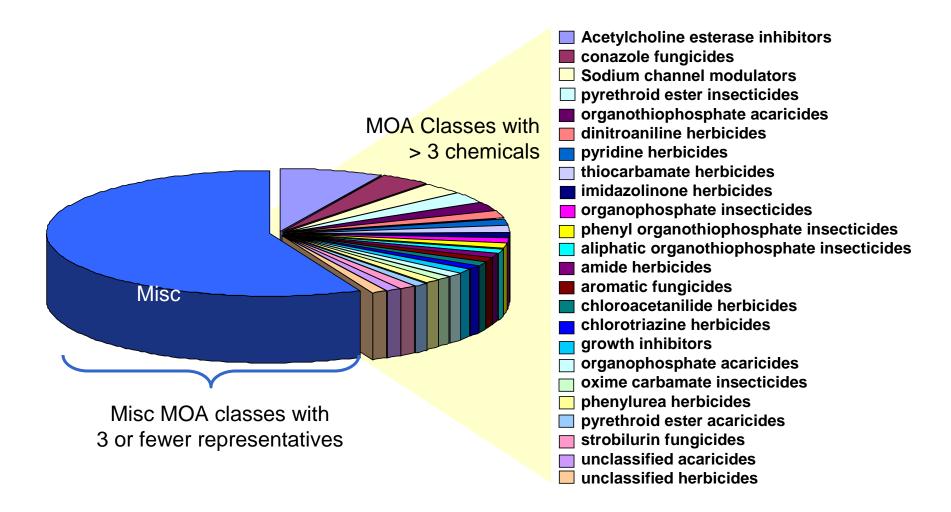
- 309 Unique Structures
- Replicates for QC
- 291 Pesticide Actives
- 9 Industrial Chemicals
- 8 Metabolites
- 56/73 Proposed Tier 1EDSP
- 14 HPV
- 11 HPV Challenge



- CHLORINE
- ORGANOPHOSPHORUS
- AMIDE
- ESTER
- ETHER
- PYRIDINE
- FLUORINE
- FLOORINE
- CARBOXYLIC ACID
- PHENOXY
- KETONE
- TRIAZINE
- CARBAMATE
- PHOSPHOROTHIOATE
- PYRIMIDINE
- BENZENE
- ORGANOCHLORINE
- AMINE
- PYRETHROID
- ☐ SULFONYLUREA
- □ TRIAZOLE
- UREA
- IMIDAZOLE
- NITRILE
- ALCOHOL
- CYCLO
- PHOSPHORODITHIOATE
- THIOCARBAMATE
- ANILINE
- THIAZOLE
- DINITROANILINE
- OXAZOLE
- PHOSPHATE
- IMINE
- NITRO
- PHENOL
- PHTHALIMIDE
- PYRAZOLE 13
- SULFONAMIDE



Mode-of-Action Classes in ToxCast_320





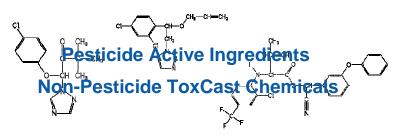
The Toxicity Data

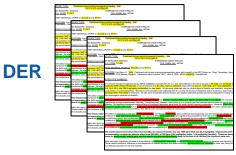


ToxRefDB: Capturing the Legacy Data

 Chemical & Toxicity Coverage

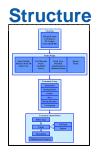
Source Data



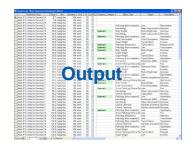


Database

ToxRefDB







Applications







Extraction of DER information

Agency

DP BARCODE: D257223 P.C. CODE: 111901

TEST MATERIAL (PURITY): Imazalil (purity ≥97.4%)

SYNONYMS: R023979

Van Deun, K., 1999. Combined oral chronic toxicity/carcinogenicity study with Imazalil in the SPF Wistar rat. Dept. Toxicology, Janssen Res CITATION:

2340 Beerse, Belgium, Laboratory report number, 3817, June 8, 1999, MRID 44858001, Unpublished

SPONSOR: Janssen Pharmaceutica N.V., 2340 Beerse, Belgium

EXECUTIVE SUMMARY:

STUDY TYPE: Combined chronic toxicity/oncogenicity feeding - Rat OPPTS 870,4300 [683-5]

SUBMISSION CODE: S564270 TOX. CHEM. NO.: 497AB

Treatment Group Info

> Treatmentrelated Effects

Chemical Info

Study Design

Endpoint/ **Critical Effects**

In a chronic toxicity/oncogenicity study (MRID 44858001), Imazalil (≥97.4% a.i.) was administered in the diet to groups of 50 male and 50 female Hannover substrain (SPF) Wistar-derived rats at concentrations of 0, 50, 200, 1200, or 2400 ppm (equivalent to 0.0, 2.7, 10.8, 65.8, and 134.8 mg/kg/day for males and 0.0, 3.6, 14.6, 85.2, and 168.8 mg/kg/day for females) for two years. All rats were observed daily for clinical signs of toxicity and morbidity, weighed weekly, and food consumption monitored biweekly. Blood and urine samples were collected after 6, 12, and 18 months of treatment and at study end. Surviving rats were sacrificed after 104 weeks of treatment. All rats were necropsied and the tissues and organs inspected grossly and microscopically for toxicity-related effects and the carcinogenic potential of Imazalil.

The absolute weights of most organs were decreased while their weights relative to body weight increased for male and female rats in the 1200 and 2400 ppm treatment groups. These effects are considered related to inanition and inappetence and not a direct result of Imazalil treatment. However, effects found in the liver and thyroid was considered directly related to treatment. The absolute liver weight of male rats in the 2400 ppm group was increased while it was decreased in female rats. The associated relative liver veights of male and female rats in the 1200 and 2400 ppm groups were significantly increased 9-26%. In addition, the absolute and relative thyroid weights of male but not female ats in the 1200 and 2400 ppm groups were increased.

The effect of treatment on the liver (males and females) and thyroid (males only) were confirmed microscopically, but had distinct sex-related etiologies. The incidence of clear cell and basophilic foci was equivocal while cosinophilic foci were significantly increased for male rats in the 2400 ppm group. In female rats of the 2400 ppm group, the incidences of clear cell and basophilic foci were significantly decreased but the incidence of eosinophilic foci was unaffected. Also, the incidence of hepatocyte falty vacuolation was increased only in male rats of the 1200 ppm and 2400 ppm groups while the incidence of pigmentation was increased only in females of the 200, 1200, and 2400 ppm groups. In addition, the location of hepatocellular hypertrophy was distinctly different. Female rats in the 1200 and 2400 ppm groups had significant increases in centriacinar and periacinar hypertrophy while male rats only had centriacinar hypertrophy. Finally, the incidence of thyroid follicular cell hyperplasia was increased only in male rats of the 1200 ind 2400 ppm groups.

The lowest observed adverse effect level (LOAEL) for male and female rats was 1200 pp m (65.8 and 85.2 mg/kg/day, respectively) with a corresponding no observed adverse effect level (NOAEL) of 200 pp m (10.8 mg/kg/day males, 14.6 mg/kg/day females). These are based on the effects found on body weight, weight gain, and the macro- and microscopic effects noted in the liver of all rats and the thyroid of male rats.

Wale rats had a significant increase in the incidence of hepatocellular adenomas and thyroid follicular neoplasia while no increase was found for female rats. These results indicat difference in the disposition of Imazalil between the sexes increases hepatic and thyroid neoplasia in male rats, likely through differences in metabolic activation of the test naterial

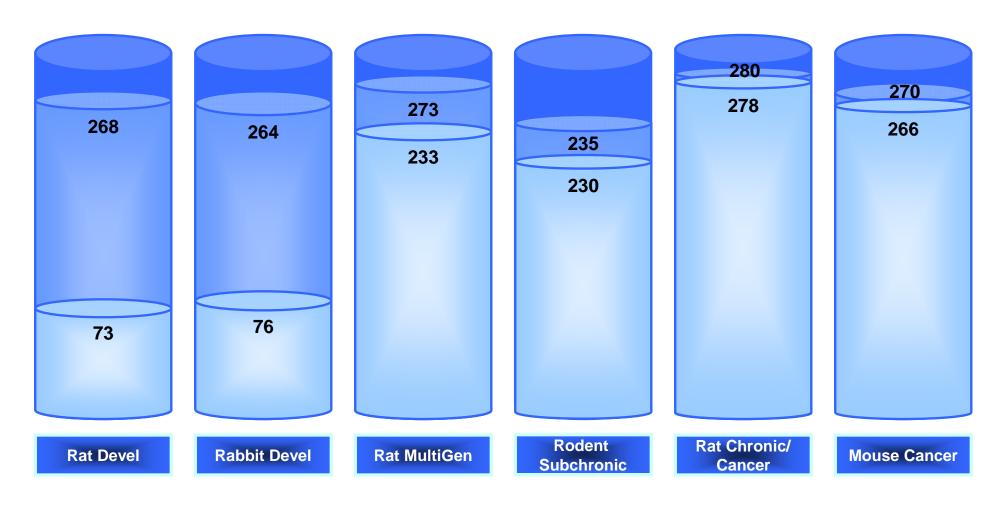
17

This chronic toxicity/oncogenicity study in the rat is Acceptable/guideline and satisfies the guideline requirement for a combined chronic toxicity/oncogenicity study in rats [83-5]. No deficiencies were noted for this study.



Data Entry Status

ToxCast Phase I Chemicals Only Total: 291 Pesticides



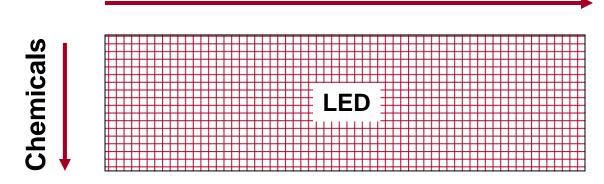


Chemical Classification

- Created directly from ToxRefDB output
- Combination of:
 - Chemical (CAS No. & Chemical Name)
 - Effect
 - Study Type (Chronic, Multigeneration, Developmental)
 - **Species** (Rat, Mouse, Rabbit)
 - Effect Type (Body weight, Clinical Chemistry, Hematology, Pathology, etc.)
 - Effect Target (Target-organ, ALT, AST, ALP, Cholesterol, etc.)
 - Effect Description (Increase, Decrease, Hypertrophy, Hyperplasia, Adenoma)
 - Aggregated Effects Collection of related individual effects
 - LED (Lowest Effective Dose)
 - Not used as regulatory term
 - · Minimum dose of observed effect



(StudyType_Species_EffectType_Target_Description)





Summary Statistics

Subchronic Rat, Chronic/Cancer Rat, and Cancer Mouse Studies

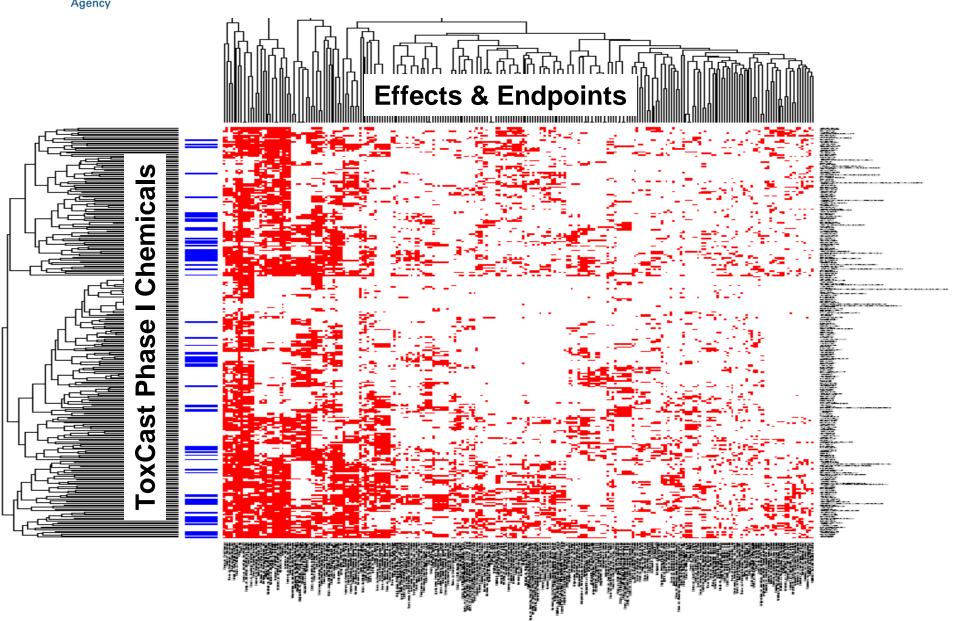
	Chemicals	Studies	Treatment Groups	Treatment Groups with Effects	Effects ^a	Critical Effects ^b	
Total	334	831	9,466	4,431	31,427	4,865	
Subchronic Rat	236	251	2,179	1,370	11,796	1,739	
Chronic/Cancer Rat	281	300	4,228	1,721	12,215	1,822	
Cancer Mouse	266	280	3,059	1,340	7,416	1,304	

- (a) Total number of effect type, target, and description combinations assigned to any treatment group
- (b) Effects that are criteria for establishing the systemic LOAEL

>9,000 Treatment groups
46% with at least one effect
31,000 Total effects
~5,000 Total Effects at the LED
~1,300 Different types of effects

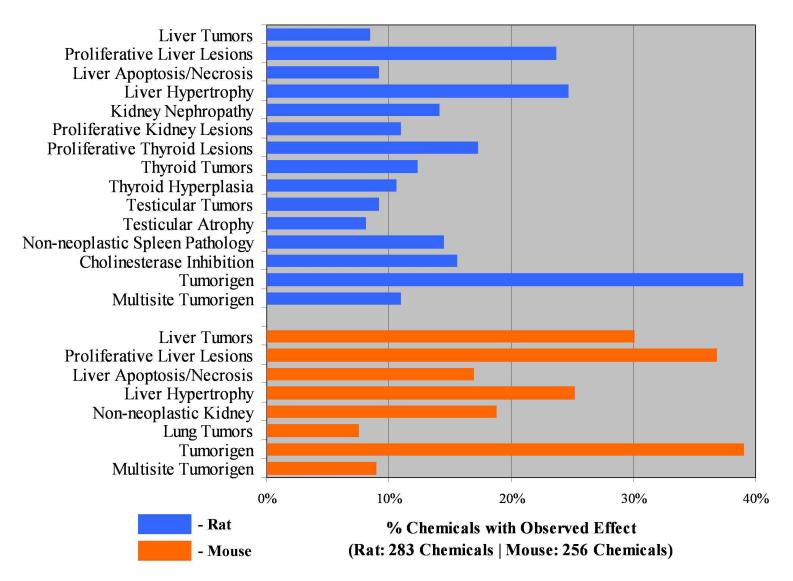


\$400 Million Dollars Worth of *In Vivo* Chronic/Cancer Bioassay Effects and Endpoints





Common Phenotypes in Chronic Rodent Studies





The HTS Assays and Toxicity Pathways

ToxCast Contracts for Generating HTS and Genomics Cell

Receptors, Enzymes



BioSeek

Metabolic Company
Transformation





Signaling







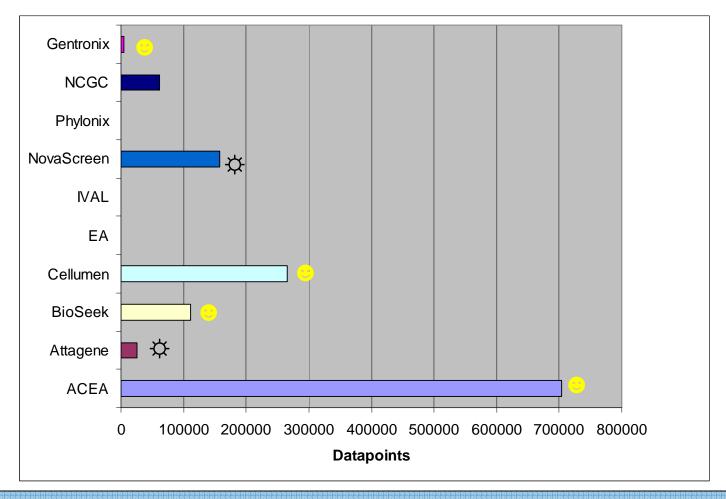




Nine contracts processing Cell ocurement; hundreds to the contract processing Cell ocurement; hundreds to the



The Deluge of Data has Started......

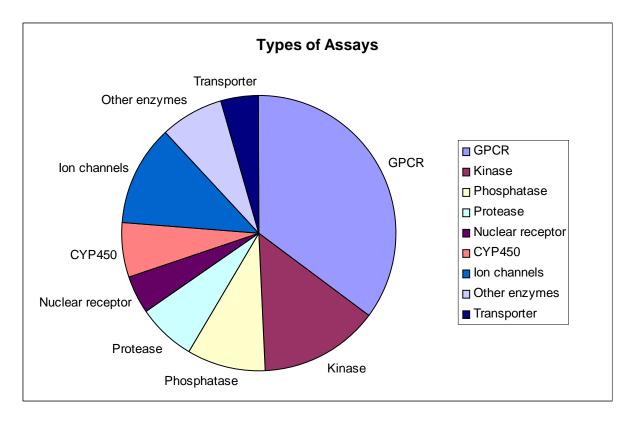


Data acquisition completed; © Concentration response follow up underway



ToxCast Biochemical Assays

- NovaScreen/Caliper LifeSciences
- 200+ Assays
- Variety of assay technologies:
 - Receptor binding
 - -Enzyme inhibition
 - -Microfluidic
 - -Fluorescent
 - Radioligand
 - -Colorimetric





Transporter

GPCR

Enzyme, other

Ion channel

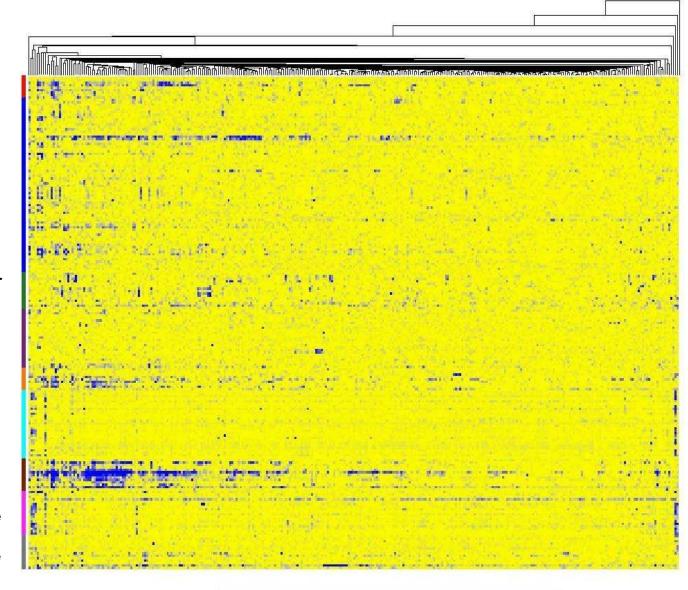
NR

Kinase

CYP450

Phosphatase

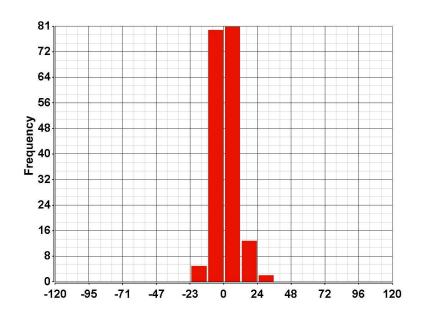
Protease

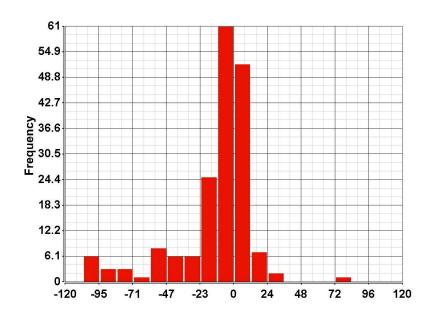


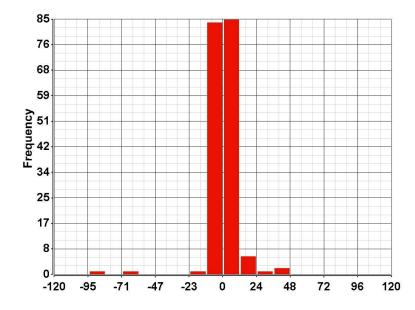
Activity (% of Control)

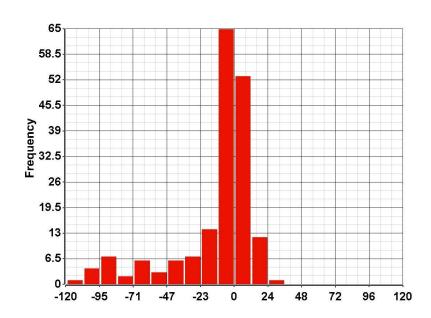


Examples of Chemical Activity Patterns



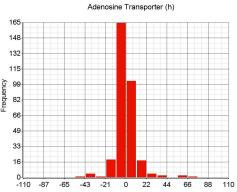


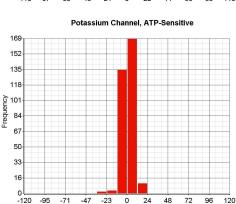


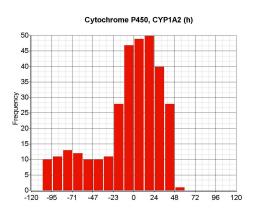


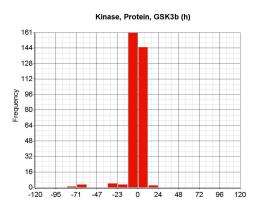


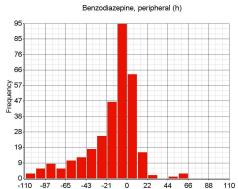
Example Biochemical Assay Results

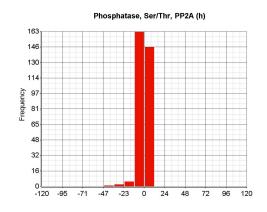


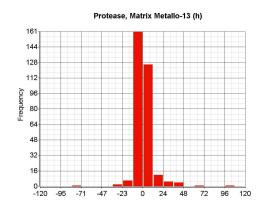


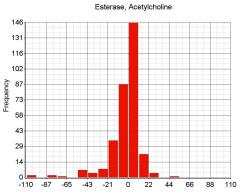


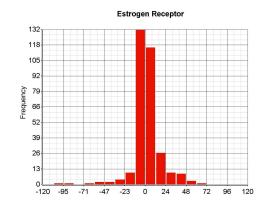


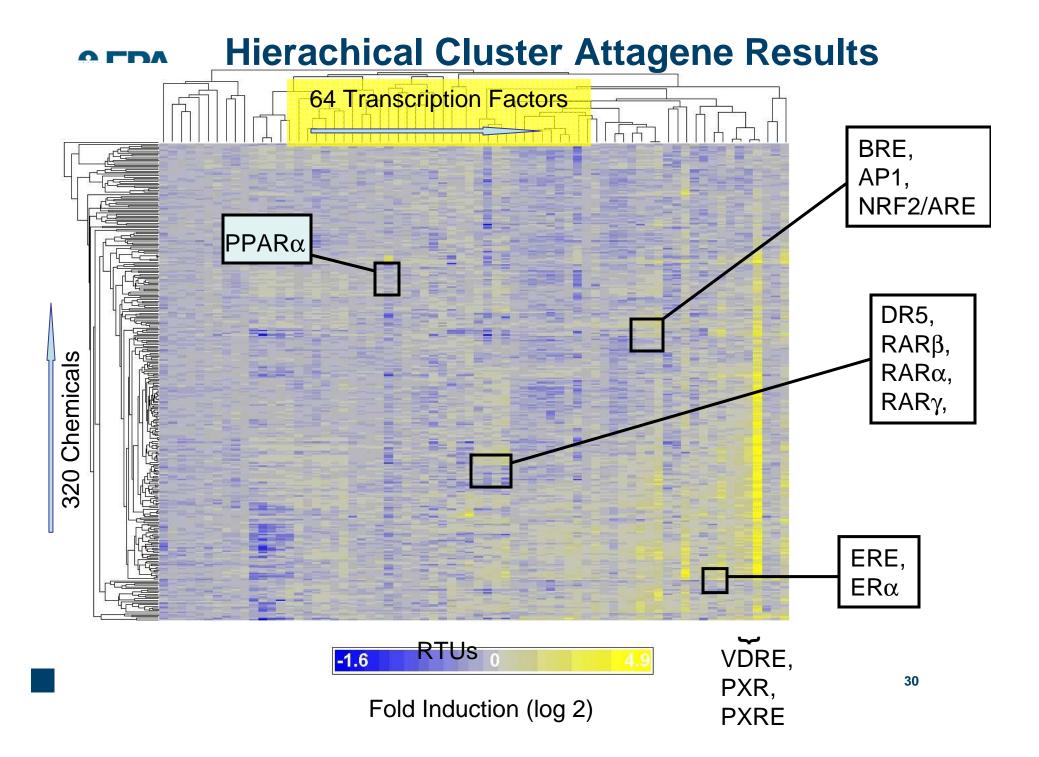














Evolution of Phase I

- ToxCast 1.0 (April, 2007)
 - Enzyme inhibition/receptor binding HTS (Novascreen)
 - NR/transcription factors (Attagene, NCGC)
 - Cellular impedance (ACEA)
 - Complex cell interactions (BioSeek)
 - Hepatocelluar HCS (Cellumen)
 - Hepatic, renal and airway cytotoxicity (IVAL)
 - In vitro hepatogenomics (IVAL, Expression Analysis)
 - Zebrafish developmental toxicity (Phylonix)
- ToxCast 1.1 (January, 2008)
 - Neurite outgrowth HCS (NHEERL)
 - Cell proliferation (NHEERL)
 - Zebrafish developmental toxicity (NHEERL)
- ToxCast 1.2 (March, 2008)
 - Organ culture: liver, kidney, lung (Hamner Institutes)
 - HTS Genotoxicity (Gentronix)
 - Toxicity and signaling pathways (Invitrogen)
 - NR Activation and translocation (CellzDirect)
 - 3D Cellular microarray with metabolism (Solidus)

8 Assay Sources & 412 Endpoints

+3 Assay Sources & 16 Endpoints

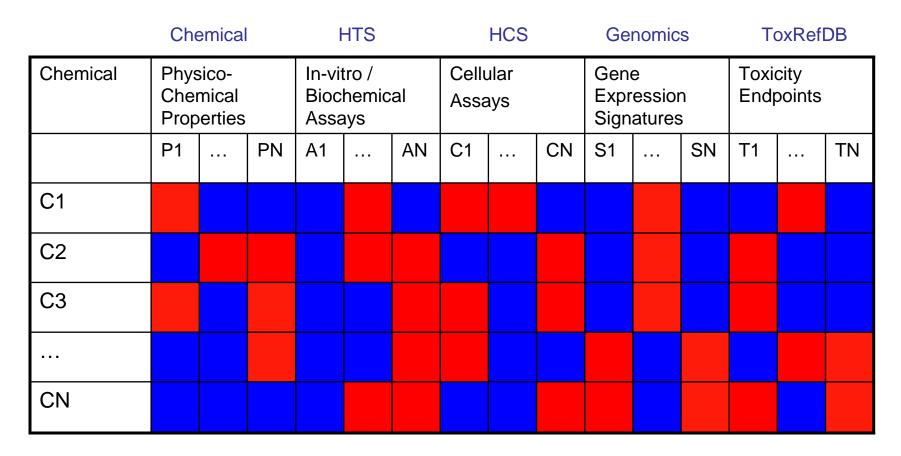
+5 Assay Sources & 32 Endpoints



Analysis and Interpretation



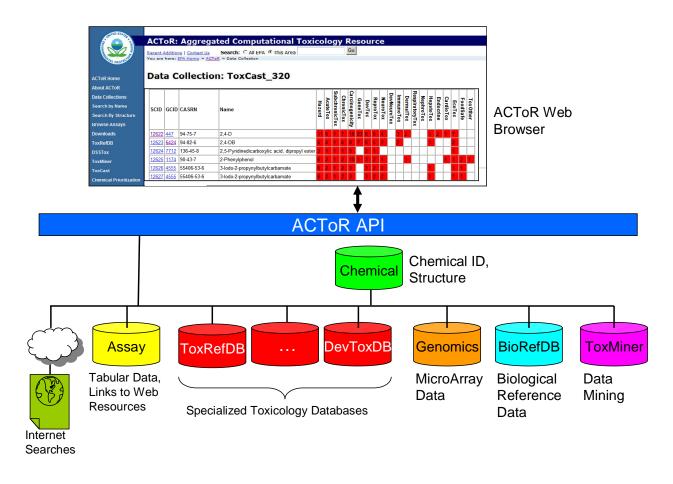
ToxCast Phase I Data Matrix



Data is both quantitative and categorical

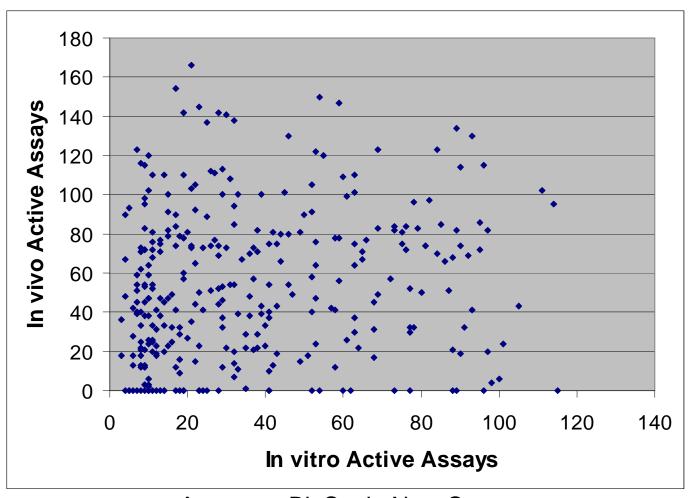


ACToR: Aggregated Computational Toxicology Resource





Does Overall In Vitro Activity Signify In Vivo Toxicity?



Attagene, BioSeek, NovaScreen



Comparing Activities by Chemical Class

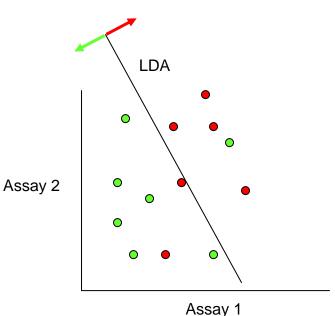
One Chemical Class vs. NovaScreen Assays

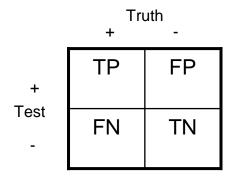
NAME	CYP2C19	CYP2C9	CYP3A1	Dopamine Transporter (Human)	CYP2D2	Androgen Receptor	Dopamine Transporter (Rat)	CYP2B6	CYP2D1	CYP3A4	Progesterone Receptor	Benzodiazepine Receptor
Chemical 1	1	1	1	1	1	0	1	0	0		0	0
Chemical 2	1	1	1	1	1	0	0	1	1	0	0	0
Chemical 3	1	1	1	0	1	0	0	0	1	1	1	0
Chemical 4	1	1	0	0	0	0	0	0	0		0	0
	1	1	1	0	1	1	0	1	1	NA	1	1
	1	1	1	1	1	0	1	1	1	NA	1	0
	1	1	1	1	1	1	1	1	1	1	1	1
	1	1	1	1	0	0	0	0	0	NA	0	0
	1	0	1	1	0	1	1	0	1	1	0	0
	1	1	1	1	1	1	1	1	1	NA	1	1
	1	1	1	0	0	0	0	1		NA	0	1
	1	1	1	0	1	1	0	1	0		1	0
	1	1	0	1	1	1	1	0	0		0	1
	1	0	0	1	0	1	1	0	0	0	0	0
	1	1	1	1	1	1	0	1	1	1	1	1
Chemical 16	1	1	1	1	0	1	1	0		NA	0	0
Totals	16	14	13	11	10	9	8	8	8	8	7	6



Association Analysis / Signatures

- Use Machine Learning methods
 - SLR: Stepwise Logistic Regression
 - LDA: Linear Discriminant Analysis
 - SVM: Support Vector Machines
 - Many others
- For each binary endpoint, build models of form
 - *Predictor* = F(assay values)
 - If
 - Predictor for a chemical meets criteria
 - Then
 - Predict endpoint to be positive for the chemical







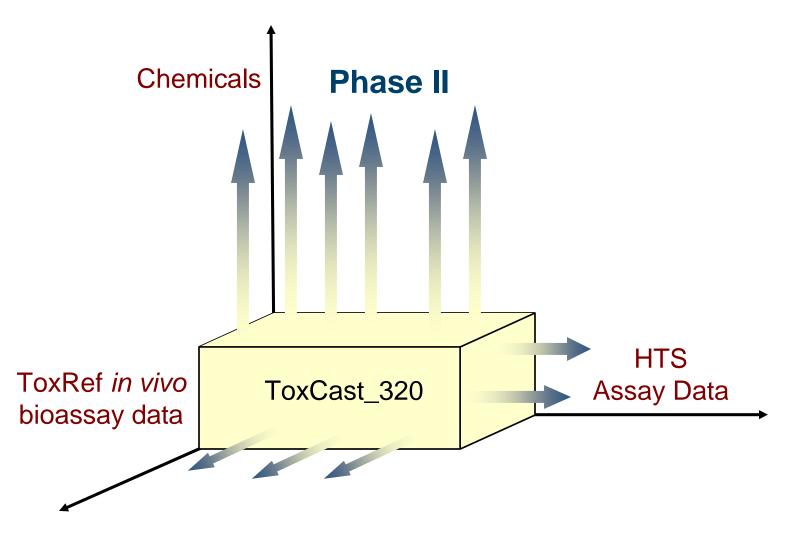
Pesticide MOA as Endpoint

MOA	Chemicals	Positives	Sensitivity	Specificity	PPV	NPV	Accuracy
thiocarbamate herbicides	303	6	1.00	0.99	0.70	1.00	0.99
dinitroaniline herbicides	303	7	1.00	0.98	0.61	1.00	0.99
Sodium channel modulators	303	11	0.90	0.97	0.51	1.00	0.93
pyrethroid ester insecticides	303	10	0.65	0.98	0.62	0.99	0.81
conazole fungicides	303	13	0.65	0.97	0.52	0.99	0.81
pyridine herbicides	303	6	0.60	0.99	0.67	0.99	0.79
Sodium channel modulators	303	11	0.60	0.98	0.53	0.99	0.79
conazole fungicides	303	13	0.50	0.98	0.53	0.98	0.74
Acetylcholine esterase inhibitors	303	27	0.50	0.97	0.66	0.96	0.74
Acetylcholine esterase inhibitors	303	27	0.52	0.95	0.57	0.96	0.73
pyrethroid ester insecticides	303	10	0.50	0.95	0.32	0.98	0.73
organothiophosphate acaricides	303	9	0.00	1.00	0.00	0.98	0.50
organothiophosphate acaricides	303	9	0.00	1.00	0.00	0.98	0.50
pyridine herbicides	303	6	0.00	1.00	0.00	0.98	0.50
thiocarbamate herbicides	303	6	0.00	1.00	0.00	0.98	0.50
dinitroaniline herbicides	303	7	0.00	0.99	0.00	0.98	0.49

Input variables: NovaScreen, Attagene, Bioseek and physical chemical properties



Beyond the Proof of Concept





Moving Forward

- Completion of Data Acquisition and Data Mining for Phase I
- Publication and Public Release of all Data
- OECD Molecular Screening Initiative
- Data Summit, Fall 2008
- Enlargement
 - Assay partners testing ToxCast_320
 - Analysis partners of Phase I data
 - Resource partners could help expand Phase II
- Related Talks
 - David Dix; Use of Genomics Data at the U.S. EPA for Predictive and Mechanistic Toxicology (Abstract #596), Monday March 17, 3:45pm, Room 611
 - Keith Houck; EPA's ToxCast Program for Predicting Hazard and Prioritizing Toxicity Testing of Environmental Chemicals, Wednesday March 19, 2:39pm, Room 6A.